

In examining what behaviors and characteristics mostly likely present a diagnosis of Autism, I started by looking at summary statistics for characteristics at a high level by general category such as age (in months), gender, ethnicity, etc. I created a field for age range, putting age into five buckets.

At this point, I did not look at the individual behavior observations recorded in fields A1 through A10. The attributes of these fields are as follows.

A1-A10: Items within [Q-Chat-10](#) in which questions possible answers : “Always, Usually, Sometimes, Rarely & Never” items’ values are mapped to “1” or “0” in the dataset. For questions 1-9 (A1-A9) in Q-chat-10, if the response was Sometimes / Rarely / Never “1” is assigned to the question (A1-A9). However, for question 10 (A10), if the response was Always / Usually / Sometimes then “1” is assigned to that question. If your child scores more than 3 (Q-chat-10- score) then there is a potential ASD traits otherwise no ASD traits are observed.

I did observe statistics such as rate of Class ASD Traits (value equal to ‘Yes’ / total records), and mean value of the Q-Chat-10 score, which is the sum of questions/values A1 through A10 for the characteristics stated above.

This yielded some interesting findings.

- Males made up 69.7% of the subjects in this study. The percentage of males showing Classic ASD traits was 72.7. The mean Q-Chat-10 score for males was 5.43.
- With respect to age range, the highest average Q-Chat-10 score was 5.64 for subjects aged 27-31 months. The highest rate of Class ASD traits was 77.54% for subjects aged 22-26 months.
- With respect to ethnicity, the highest average Q-Chat-10 score was 7 for subjects designated as Native American and Pacifica. The highest rate of Class ASD traits was 100% for Native Americans. However, only 8 subjects were designated Pacifica, and 3 Native American. If we consider the idea of sample size times probability, and sample size times 1-probability should both be greater than or equal to 10, the following table is useful for looking at sub-categories of ethnicity.

| Ethnicity | Value Counts | Proportion Total | Prob ClassASD = Y | ValCount x Prob ClassASD = Y | ValCount x Prob ClassASD = N | Mean Q-Chat-10 Score |
|----------------|--------------|------------------|-------------------|------------------------------|------------------------------|----------------------|
| White European | 334 | 31.69% | 74.85% | 250 | 84 | 5.72 |
| asian | 299 | 28.37% | 70.90% | 212 | 87 | 4.98 |
| middle eastern | 188 | 17.84% | 51.06% | 96 | 92 | 4.27 |
| south asian | 60 | 5.69% | 61.67% | 37 | 23 | 5.02 |
| black | 53 | 5.03% | 73.58% | 39 | 14 | 5.72 |

| Ethnicity | Value Counts | Proportion Total | Prob ClassASD = Y | ValCount x Prob ClassASD = Y | ValCount x Prob ClassASD = N | Mean Q-Chat-10 Score |
|---------------|--------------|------------------|-------------------|------------------------------|------------------------------|----------------------|
| Hispanic | 40 | 3.80% | 75.00% | 30 | 10 | 5.60 |
| Others | 35 | 3.32% | 82.86% | 29 | 6 | 5.63 |
| Latino | 26 | 2.47% | 76.92% | 20 | 6 | 5.92 |
| mixed | 8 | 0.76% | 62.50% | 5 | 3 | 4.63 |
| Pacifica | 8 | 0.76% | 87.50% | 7 | 1 | 7.00 |
| Native Indian | 3 | 0.29% | 100.00% | 3 | 0 | 7.00 |
| | 1054 | 100.00% | | | | |

At this point, I use SQL to create binary, “dummy” variables for all categorical values. I used the .corr() method in Python to create a table of correlation coefficients. This yielded some interesting results as well.

The correlation coefficient for male/gender to ClassicASDTraits was 0.1177.

The most significant correlation coefficient for age range to ClassicASDTraits was -0.1313 for the 12-16 month age range.

The most significant correlation coefficient for ethnicity to ClassicASDTraits was -0.1815 for Middle Eastern subjects.

The correlation coefficients for Jaundice and the Completed By... categories indicated an even weaker relationship.

Correlation coefficients for behavioral observations A1-A10 indicated strongest relationships to ClassicASDTraits. The strongest of these as follows.

- A5: Does your child pretend?, 0.5633
- A6: Does your child follow where you’re looking?, 0.5694
- A7: If you or someone else in the family is visibly upset, does your child show signs of wanting to comfort them?, 0.5632
- A9: Does your child use simple gestures, e.g. wave goodbye?, 0.5773

I will examine these relationships further with machine learning and predictive modeling.